## Supervised Neural Networks for Helioseismic Ring Diagram Inversions

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#### The Solar Subsurface

- Direct imaging is impossible
- Inferences made from observed surface wave field
- Flows cause frequency shifts



HMI Dopplergram<sup>[1]</sup>

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<sup>1</sup>sdo.gsfc.nasa.gov/

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#### **Ring-Diagram Analysis**

Machine Learning Results Summary



# Tracked 15°Patch

<sup>1</sup>Basu et al. 1999 <sup>2</sup>Greer 2015

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#### Ring-Diagram Analysis

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$$u_{x}(z_{t}) = \sum_{n,\ell} c_{n,\ell}(z_{t}) U_{x}^{n,\ell}$$
$$= \boxed{\sum_{n,\ell} \int c_{n,\ell}(z_{t}) \mathcal{K}^{n,\ell}(z)} u_{x}(z) dz \qquad \text{Linear?}$$
(1)

Computationally expensive [31 CPU hrs/CR]

<sup>1</sup>Basu et al. 1999 <sup>2</sup>Greer 2015

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## Why use machine learning?

- Solves non-linear problems
- Lots of data (>700,000 observations)
- Greatly reduces computational burden
- Replace current procedure

## What's involved?

- Supervised Learning
- Preprocessing data
- Model optimization

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ML	Train &	Prediction Time [s]	RMS	E (m/s)
	$u_x$	$u_y$	$u_x$	$u_y$
Lin	< 1	< 1	8.8	7.3
Bay	< 1	< 1	8.8	7.3
DT	19	20	8.8	7.4
RF	39	42	10.5	9.0
KNN	50	35	9.0	7.7
ANN	360	194	8.6	7.3
SVR	61441	47555	8.5	7.3

For  $z_t = 10.44 \text{ Mm}$ 



Black: *u<sub>x</sub>*, Red: *u<sub>y</sub>* Dashed: Observed, Solid: ML

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#### Have we lost some helioseismic signatures? Can we detect Rossby waves (Löptien et al. 2018) in ML?



#### Did we succeed?

- ML can be used to accurately predict future observations
- Training for all depths: < 5 CPU hrs [using entire SDO obs]
- Predictions for next CR: < 1 CPU sec [old: 31 CPU hrs]

Future: Dopplergram  $\rightarrow$  Flows

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